

Sociology 63993, Exam1 February 12, 2015

Richard Williams, University of Notre Dame, <http://www3.nd.edu/~rwilliam/>

I. *True-False.* (20 points) Indicate whether the following statements are true or false. If false, briefly explain why.

1. When working with complex survey data (and using the svy: prefix in Stata) nested models should be tested via the use of incremental F tests (i.e. you should estimate the constrained and unconstrained models separately and then use an incremental F test to contrast them).

2. The closer the tolerance of a variable is to 1, the more likely it is that you will have problems with multicollinearity.

3. The most extreme outliers on Y (i.e. the cases where Y is furthest from the mean) will always have the most influence on the regression line.

4. Cohen and Cohen's Dummy Variable Adjustment technique has been discredited and should not be used under any circumstances.

5. A researcher runs the following analysis:

```
. alpha v1 v2 v3, i
```

```
Test scale = mean(unstandardized items)
```

Item	Obs	Sign	item-test correlation	item-rest correlation	average interitem covariance	alpha
v1	3975	+	0.4842	0.1522	.2360952	0.7907
v2	3975	+	0.8448	0.6473	.0374987	0.1997
v3	3975	+	0.8836	0.5602	.0342703	0.2815
Test scale					.1026214	0.6060

Based on these results, she should drop v2 from her scale.

II. *Short answer.* Discuss all three of the following problems. (15 points each, 45 points total.) In each case, the researcher has used Stata to test for a possible problem, concluded that there is a problem, and then adopted a strategy to address that problem. Explain (a) what problem the researcher was testing for, and why she concluded that there was a problem, (b) the rationale behind the solution she chose, i.e. how does it try to address the problem, and (c) one alternative solution she could have tried, and why. (NOTE: a few sentences on each point will probably suffice – you don't have to repeat everything that was in the lecture notes.)

//-1.

. reg health age weight height i.female i.race

Source	SS	df	MS	Number of obs = 800		
Model	193.406808	6	32.234468	F(6, 793)	=	28.89
Residual	884.811942	793	1.11577798	Prob > F	=	0.0000
				R-squared	=	0.1794
				Adj R-squared	=	0.1732
				Root MSE	=	1.0563
Total	1078.21875	799	1.34946026			

health	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	-.0211716	.0022829	-9.27	0.000	-.0256528	-.0166904
weight	-.0039733	.0028427	-1.40	0.163	-.0095535	.0016068
height	.0127136	.0058336	2.18	0.030	.0012625	.0241647
1.female	.1804978	.1031217	1.75	0.080	-.0219259	.3829216
race						
Black	-.7038075	.0977277	-7.20	0.000	-.895643	-.5119719
Other	-.0574042	.1235336	-0.46	0.642	-.2998958	.1850873
_cons	2.645635	1.012356	2.61	0.009	.6584201	4.632851

. sum, sep(6)

Variable	Obs	Mean	Std. Dev.	Min	Max
race	800	1.445	.7071776	1	3
age	1100	48.88364	17.46024	20	74
height	1100	167.2078	10.19798	138.5	200
weight	1100	71.0562	15.31384	30.84	149.69
health	1100	3.401818	1.172321	1	5
female	1100	.5218182	.499751	0	1

. tab1 race

-> tabulation of race

1=white, 2=black, 3=other	Freq.	Percent	Cum.
White	545	68.13	68.13
Black	154	19.25	87.38
Other	101	12.63	100.00
Total	800	100.00	

. mi set mlong

. mi register imputed race

(300 m=0 obs. now marked as incomplete)

. mi impute mlogit race health age height weight female, add(50) rseed(2232)

```

Univariate imputation                Imputations =      50
Multinomial logistic regression      added =      50
Imputed: m=1 through m=50           updated =      0
  
```

Variable	Observations per m			Total
	Complete	Incomplete	Imputed	
race	800	300	300	1100

(complete + incomplete = total; imputed is the minimum across m of the number of filled-in observations.)

. mi estimate: reg health age weight height i.female i.race

```

Multiple-imputation estimates      Imputations =      50
Linear regression                  Number of obs =    1100
                                   Average RVI    =     0.1216
                                   Largest FMI    =     0.2769
                                   Complete DF    =     1093
DF adjustment: Small sample       DF:   min    =     357.50
                                   avg      =     823.32
                                   max      =    1052.35
Model F test:   Equal FMI         F(   6, 1032.2) =     33.72
Within VCE type: OLS              Prob > F      =     0.0000
  
```

health	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
age	-.0200064	.0019955	-10.03	0.000	-.0239221 -.0160908
weight	-.0051971	.0024855	-2.09	0.037	-.0100745 -.0003198
height	.015661	.0050562	3.10	0.002	.0057391 .0255829
1.female	.2240806	.0902576	2.48	0.013	.0469683 .4011929
race					
2	-.7150266	.09896	-7.23	0.000	-.9096434 -.5204098
3	-.0666072	.1224521	-0.54	0.587	-.3074037 .1741892
_cons	2.161208	.8808568	2.45	0.014	.4326168 3.889799

//-2.

. reg y x

```

Source |      SS      df      MS      Number of obs =    100
-----+-----+-----+-----+-----
Model |  387.15257    1  387.15257  F(   1,   98) =  204.90
Residual |  185.17243   98  1.88951459  Prob > F      =   0.0000
-----+-----+-----+-----
Total |   572.325   99  5.78106061  R-squared     =   0.6765
                                   Adj R-squared  =   0.6732
                                   Root MSE      =   1.3746
  
```

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
x	1.977532	.1381521	14.31	0.000	1.703373 2.25169
_cons	1.94407	.1374596	14.14	0.000	1.671286 2.216854

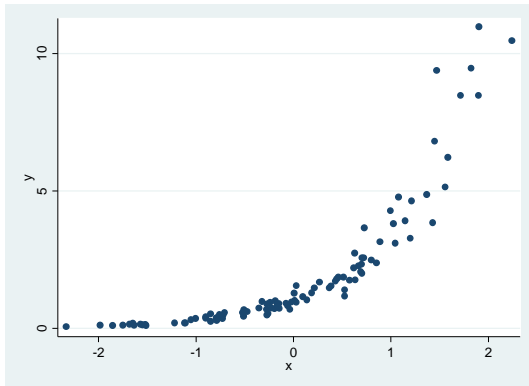
```
. estat hettest
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance
Variables: fitted values of y

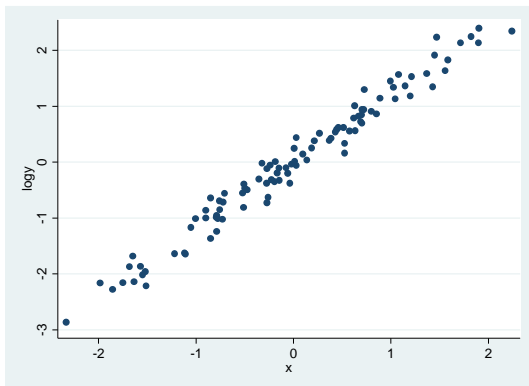
```
chi2(1)      =    32.27
Prob > chi2  =    0.0000
```

```
. twoway scatter y x, name(g1)
```



```
. gen logy = log(y)
```

```
. twoway scatter logy x, name(g2)
```



```
. reg logy x
```

Source	SS	df	MS			
Model	142.56	1	142.56	Number of obs =	100	
Residual	3.96000021	98	.040408165	F(1, 98) =	3528.00	
Total	146.520001	99	1.48000001	Prob > F =	0.0000	
				R-squared =	0.9730	
				Adj R-squared =	0.9727	
				Root MSE =	.20102	

logy	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x	1.2	.0202031	59.40	0.000	1.159908	1.240092
_cons	2.14e-09	.0201018	0.00	1.000	-.0398913	.0398913

```
. estat hettest
```

```
Breusch-Pagan / Cook-Weisberg test for heteroskedasticity  
Ho: Constant variance  
Variables: fitted values of logy
```

```
chi2(1) = 0.02  
Prob > chi2 = 0.8781
```

```
//-3.
```

```
. reg y x
```

Source	SS	df	MS	Number of obs =	3975
Model	559.504598	1	559.504598	F(1, 3973) =	0.04
Residual	57188892.5	3973	14394.3852	Prob > F =	0.8437
				R-squared =	0.0000
				Adj R-squared =	-0.0002
Total	57189452.1	3974	14390.9039	Root MSE =	119.98

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
x	.3505943	1.778278	0.20	0.844	-3.135828 3.837017
_cons	3.50558	2.474016	1.42	0.157	-1.344881 8.356041

```
. dfbeta
```

```
_dfbeta_1: dfbeta(x)
```

```
. extremes _df* y x
```

obs:	_dfbeta_1	y	x
2846.	-20.48698	7560.241	.1534038
2100.	-.0031762	-6.815401	3.137415
3828.	-.0019974	-2.850775	3.062574
70.	-.0019538	-3.89073	2.776838
3739.	-.0019023	-5.510675	2.447441

2439.	.0025574	-8.791584	-.8340001
2444.	.002686	-7.56208	-1.147358
171.	.0027336	-11.03818	-.6557877
1442.	.0027977	-7.366304	-1.281223
2546.	.0028055	-12.22433	-.5724241

```
. drop _df*
```

```
. replace y = y/1000 in 2846
```

```
(1 real change made)
```

. reg y x

Source	SS	df	MS			
Model	11237.3545	1	11237.3545	Number of obs =	3975	
Residual	64343.0002	3973	16.1950668	F(1, 3973) =	693.88	
				Prob > F =	0.0000	
				R-squared =	0.1487	
				Adj R-squared =	0.1485	
Total	75580.3548	3974	19.0187103	Root MSE =	4.0243	

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x	1.571213	.0596478	26.34	0.000	1.45427	1.688156
_cons	.5203355	.0829846	6.27	0.000	.3576392	.6830318

. dfbeta

_dfbeta_1: dfbeta(x)

. extremes _df* y x

obs:	_dfbeta_1	y	x
2100.	-.1018125	-6.815401	3.137415
619.	-.0739862	12.73301	-.6295367
2606.	-.0678742	12.7148	-.5239393
3828.	-.065618	-2.850775	3.062574
2574.	-.0643436	10.71742	-.6628681

1111.	.0597705	13.48404	2.771187
171.	.0599936	-11.03818	-.6557877
2724.	.0613352	13.84974	2.727732
665.	.0637711	13.70689	2.887728
2546.	.0638679	-12.22433	-.5724241

III. *Computation and interpretation.* (35 points total) The Center for Disease Control is very concerned about the anti-vaccination movement in the United States. According to the World Health Organization (<http://www.who.int/mediacentre/factsheets/fs286/en/>), measles is one of the leading causes of death among young children worldwide even though a safe and cost-effective vaccine is available. In the United States, the number of measles cases has skyrocketed in recent years, largely because growing numbers of parents are choosing not to vaccinate their children. Various explanations have been offered.

- Due to a now discredited study (<http://www.newsweek.com/autism-how-childhood-vaccines-became-villains-82273>), some parents fear that the measles vaccine can cause autism.
- Another recent article claimed that vaccination refusal was a “white privilege” problem: it takes money and time to refuse vaccinations, and whites are more likely to have that money and time than are minorities (<http://www.xojane.com/issues/vaccination-refusal-white-privilege>).
- Finally, another study, recently reported on NPR (<http://www.npr.org/blogs/health/2015/02/06/384322665/to-get-parents-to-vaccinate-their-kids-dont-ask-just-tell>), claims that a doctor’s approach has a major impact on

whether or not parents vaccinate their children. When doctors just simply presumed that the parent was going to be fine with the vaccines that the doctor was going to recommend (e.g. “Johnny is due for his DTaP shot today”), parents were much more likely to get their child vaccinated than they were when the doctor asked them how they felt about vaccination.

To assess the validity and importance of these different claims, the CDC has collected complete data from 2000 parents of young children. The items included in the survey are:

Variable	Description
vaccination	Scale that measures feelings about vaccination. Ranges from 0 = extremely negative about vaccinations to 100 = extremely positive. This is the dependent variable.
white	Coded 1 if white, 0 if non-white
autism	Scale that measures beliefs about whether vaccines can lead to autism. 0 = no chance that vaccinations can cause autism to 100 = extremely likely that vaccinations can cause autism.
approach	Scale that measures how forceful the children’s doctor is in pushing vaccinations. 0 = not forceful at all, 100 = just assumes the parent will want their child vaccinated.

An analysis of the data yields the following results. [NOTE: You’ll need some parts of the following to answer the questions, but other parts are extraneous. You’ll have to figure out which is which.]

```
. sum vaccinate white autism approach
```

Variable	Obs	Mean	Std. Dev.	Min	Max
vaccinate	2000	47.3155	26.8503	1	100
white	2000	.885	.3191017	0	1
autism	2000	27.4605	17.03163	0	71
approach	2000	60.605	15.83222	0	100

```
. reg vaccinate i.white autism approach, vce(robust)
```

```
Linear regression                               Number of obs =    2000
                                                F(   3, 1996) =   47.22
                                                Prob > F       =   0.0000
                                                R-squared      =   0.0648
                                                Root MSE      =  25.985
```

	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
vaccinate						
white						
White	-4.363207	1.798159	-2.43	0.015	-7.889673	-.8367423
autism	-.2537355	.0352135	-7.21	0.000	-.3227946	-.1846764
approach	.2573966	.0386828	6.65	0.000	.1815336	.3332596
_cons	42.54512	3.013613	14.12	0.000	36.63497	48.45528

. reg vaccinate i.white autism approach

Source	SS	df	MS	Number of obs =	2000
Model	93383.6877	[1]	31127.8959	F(3, 1996) =	[2]
Residual	1347772.23	1996	675.236589	Prob > F =	0.0000
Total	1441155.92	1999	[4]	R-squared =	[3]
				Adj R-squared =	0.0634
				Root MSE =	25.985

vaccinate	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
white						
White	-4.363207	1.838179	[5]	0.018	-7.968157	-.7582576
autism	-.2537355	.0357901	-7.09	0.000	-.3239253	-.1835457
approach	.2573966	.0387209	6.65	0.000	.1814591	.3333342
_cons	42.54512	3.099787	13.73	0.000	36.46596	48.62428

. estat hettest

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity
 Ho: Constant variance
 Variables: fitted values of vaccinate
 chi2(1) = 2.40
 Prob > chi2 = 0.1212

. estat imtest

Cameron & Trivedi's decomposition of IM-test

Source	chi2	df	p
Heteroskedasticity	14.88	8	0.0614
Skewness	44.28	3	0.0000
Kurtosis	56.84	1	0.0000
Total	116.00	12	0.0000

. testparm i.white autism approach

(1) 1.white = 0
 (2) autism = 0
 (3) approach = 0
 F(3, 1996) = 46.10
 Prob > F = 0.0000

. test approach = -autism

(1) autism + approach = 0
 F(1, 1996) = 0.00
 Prob > F = 0.9514


```
. pcorr vaccinate white autism approach
(obs=2000)
```

Partial and semipartial correlations of vaccinate with

Variable	Partial Corr.	Semipartial Corr.	Partial Corr.^2	Semipartial Corr.^2	Significance Value
white	-0.0531	-0.0514	0.0028	0.0026	0.0177
autism	-0.1567	-0.1535	0.0246	0.0235	0.0000
approach	0.1472	0.1439	0.0217	0.0207	0.0000

```
. reg vaccinate i.white autism approach i.white#i.white
```

Source	SS	df	MS	Number of obs =	2000
Model	93383.6877	3	31127.8959	F(3, 1996) =	46.10
Residual	1347772.23	1996	675.236589	Prob > F =	0.0000
Total	1441155.92	1999	720.938429	R-squared =	0.0648
				Adj R-squared =	0.0634
				Root MSE =	25.985

vaccinate	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
white					
White	-4.363207	1.838179	-2.37	0.018	-7.968157 - .7582576
autism	-.2537355	.0357901	-7.09	0.000	-.3239253 - .1835457
approach	.2573966	.0387209	6.65	0.000	.1814591 .3333342
_cons	42.54512	3.099787	13.73	0.000	36.46596 48.62428

a) (10 pts) Fill in the missing quantities [1] – [5]. (A few other values may have also been blanked out, but you don't need to fill them in.)

b) (25 points) Answer the following questions about the analysis and the results, explaining how the printout supports your conclusions.

1. Summarize the key findings. In your discussion, indicate whether or not the beliefs that caused the CDC to examine the variables in the first place were borne out by the results.

2. An additional 227 cases were dropped from the analysis because they were missing data on race and/or approach. If you wanted to keep those cases in the analysis, what multiple imputation method or methods would you recommend using (e.g. logit, mlogit, regress, ologit, pmm, poisson, or something else)? Briefly explain why.

3. The researchers ran the regression with vce(robust) and then again without vce(robust). They noticed that the coefficients did not change, so they decided to not use vce(robust). Do you think this was sound reasoning on their part? Whether it was or was not sound reasoning is there other evidence from the printout that supports or challenges their decision to not use robust standard errors?

4. Some of the researchers believe that beliefs about autism have the greatest impact on support for vaccinations. Others say that it is the doctor's approach that matters the most. Still

others contend that both variables are about equally important and that the differences in their effects are either trivial or non-existent. What is your own position on this, and why? Be sure to cite multiple pieces of information from the printout to support your position.

5. An undergraduate intern has been told that it is often important to include squared terms in models, so he added white^2 to the final regression. To his surprise, none of the results changed. Indeed the squared term didn't even show up in the output. Explain to him why this was the case. [Note: You can draw on your vast sociological expertise in offering a theoretical explanation for this. Or, if that student happens to be visiting us this weekend, you can explain why Notre Dame Sociology may want to think twice before admitting him.]